Managing Data



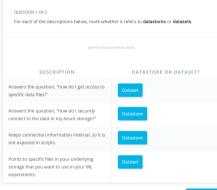
As we just discussed, Azure Machine Learning has two data management tools that we need to consider: **Datastores** and **datasets**. At first the distinction between the two may not be entirely clear, so let's have a closer look at what each one does and how they are related.

Datastores vs. Datasets

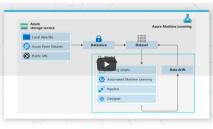


Datastores offer a layer of abstraction over the supported Azure storage services. They store all the information needed to connect to a particular storage service. Datastores provide an access mechanism that is independent of the computer resource that is used to drive a machine learning process.

Datasets are resources for exploring, transforming, and managing data in Azure ML. A dataset is essentially a reference that points to the data in storage. It is used to get specific data files in the datastores.



The Data Access Workflow



- 1. Create a datastore so that you can access storage services in Azure.
 2. Create a dataset, which you will subsequently use for model training in your machine learning experiment.
 3. Create a dataset monitor to detect issues in the data, such as data drift.

In the video, we mentioned the concept of data drift. Over time, the input data that you are feeding into your model is likely to change—and this is what we mean by data drift. Data drift can be problematic for model accuracy. Since you trained the model on a certain set of data, it can become increasingly inaccurate and the data changes more and more over time. For example, if you train a model to detect spam in email, if, may become less accurate as new types of spam arise that are different from the spam on which the model was trained.

As we noted in the video, you can set up dataset manitors to detect data drift and other issues in your data. When data drift is detected, you can have the system automatically update the input dataset so that you can retrain the model and maintain its accuracy.

